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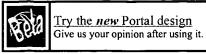
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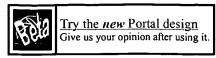
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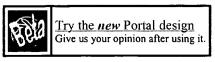
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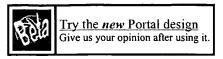
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Paul Suetens , Pascal Fua , Andrew J. Hanson ACM Computing Surveys (CSUR) March 1992

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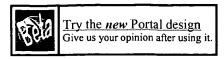
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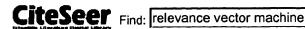


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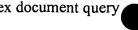
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18 5. Hyperparameter Inference

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